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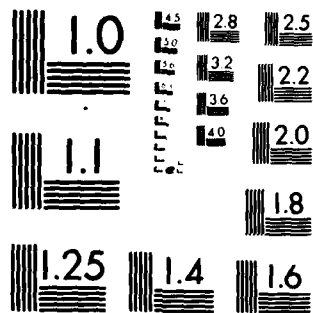
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ABSTRACT

The algorithm for synthesizing relational data base schema in the 3rd normal form assumes uniqueness of functional dependencies. This assumption is examined and a method for checking the assumption is presented.

I. INTRODUCTION

Beeri and Bernstein [2] have developed a fast algorithm for synthesizing relational data base schema in the 3rd normal form from a given set of FDs such that the resulting schema embodies the original FD's. They take an axiomatic approach to FDs and use a set of axiom schema to derive all the FDs that follow from a given set of FDs. If G is a given set of FDs, G^+ denotes the set of derived FDs. In order for their algorithm to work they must make the following "uniqueness" assumption.

Let X be a set of attributes and A an attribute. If $X \twoheadrightarrow A$ G^+ then any derivation of $X \twoheadrightarrow A$ represents the same "user intent".

In [2a] Beeri and Bernstein achieve the uniqueness assumption by assuming that all the relations of a data base are projections of a "universal relation". Since a universal relation never "really" exists the verification of the uniqueness assumption remains a problem.

When Beeri and Bernstein's algorithm is used to synthesize relational data base schema an attempt should be made to verify the uniqueness assumption. Even though Beeri and Bernstein's algorithm is linear any attempt to verify the uniqueness assumption is doomed to exponential time.

An automatic checking for violations of the uniqueness assumption is preferable to interactively "showing" each derivation to the user. Such a semantic analyzer is difficult to find since it is not known how to formalize the "user intent" of an FD. As a partial solution we classify FDs into three types, regular, injective and computable. Armstrong's [1] axioms can be applied to these types so that every derivation of an FD will result in classifying the FD as one of the three types (given the types of the initial FDs). When two derivations of an FD result in two different classifications then we have a violation. If two derivations both result in a calculable FD it is sometimes possible to decide that the calculations are different and there is a violation. In other cases it would not be known if there was a violation.

The usual solution to a violation of uniqueness is to rename some attributes. We show that this may lead to "problems" and that sometimes certain derivations must be "outlawed."

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II.3 Preliminaries

We assume the reader is familiar with the notation and results in Beeri and Bernstein [2]. Our view of relational data bases is somewhat similar to that of Cadiou [4] and Nicholas [8]. There are two notions of a relation; intension and extension.

The intention of a relation should include as much of the "user intent" as possible. The intention of a relation consists of:

1. $R(A_1, \dots, A_n)$ - a relational form, where R is the name of the relation and $\{A_1, \dots, A_n\}$ are the attributes

2. A set of keys.

Remark: 1. and 2. are usually called a relation scheme and in Beeri and Bernstein [2] this is all that is meant by the "intention."

3. Functional dependencies
4. Other types of dependencies
5. Domain definitions of the attributes
6. Other integrity constraints (See Eswarian [5] and Hammer Mcleod [7] for a taxonomy).

For each intension R there are many extensions. Each extension (or instance) of R is a finite set, R , of n -tuples satisfying the constraints of the intention.

The intension of a data base is a finite collection of relational intensions with additional integrity constraints (that include more than one relation). Attributes and their domain definitions are invariant over the data base so the FDs (and other dependencies) can be considered to reside in the data base as a whole.

The constraints on a data base can be stated in any appropriate language such as: first order predicate logic, SEQUEL, QUERY BY EXAMPLE. It is possible to discuss the set of all extensions of a data base but many questions (consistency, derivability) may be undecidable. For details consult Gallaire and Minker [6] and Nicholas [8].

The constraints for which the above questions are important are those that affect the structure of the data base. FDs affect the relational data base schema since the normal forms are stated in terms of the FDs. Fortunately, under the uniqueness assumption, questions of consistency and derivability about FDs are decidable.

We shall review some notation from Beeri and Bernstein [2] and describe their program. An FD is denoted by $X \twoheadrightarrow Y$, where X and Y are sets of attributes. The only information that the above notation imparts is that for any relation R whose attributes include $X \cup Y$ and for any instance r of R , if two tuples coincide on X they must also coincide on Y . Sometimes $f: X \twoheadrightarrow Y$ is written, where f denotes a canonical name for the partial function from $\text{dom}(X)$ to $\text{dom}(Y)$ which is dependent on the extension (and changes

as the extension does).

Armstrong [1] gave a set of axiom schema for deriving FDs from a given set of FDs and shows that the system is sound and complete. Beeri and Bernstein use the following equivalent axioms.

A_1 : (Reflexivity) $X \twoheadrightarrow X$

A_2 : (Augmentation) If $X \twoheadrightarrow Z$ then $X \cup Y \twoheadrightarrow Z$

A_3 : (Pseudotransitivity) If $X \twoheadrightarrow Y$ and $Y \cup Z \twoheadrightarrow W$ then $X \cup Z \twoheadrightarrow W$.

If G is a set of FDs, then G^+ is the closure of G under the above axioms. An important part of Beeri and Bernstein's algorithm is to compute G^+ . If $X \twoheadrightarrow Y$ can be derived from G it can be derived by an infinite number of derivations. By the uniqueness assumption Beeri and Bernstein can assume any derivation of $X \twoheadrightarrow Y$ represents a unique "user intent." Hence they need only search for one such derivation. They only have to search derivation trees of height at most the number of attributes among the G since a derivation with a loop (i.e. one that goes through an attribute twice) is the same without the loop since $X \twoheadrightarrow X$ must be the identity mapping by uniqueness.

III. Checking and Correcting the Uniqueness Assumption

Any method for verifying the uniqueness assumption will involve comparing different derivations of a single FD. By the above method, if $X \twoheadrightarrow Y$ is not unique there are two derivations of $X \twoheadrightarrow Y$ by trees of at most height twice the number of attributes

(since at most one loop is needed). Hence if we had an algorithm for deciding whether two derivations represent the same "user intent" the uniqueness problem would be decidable. Since we would have to search all derivations the solution is at best exponential. When a violation of uniqueness is discovered the usual remedy is to change attribute names so that two different FDs are produced. This would change the final relational database schema synthesized by the Beeri and Bernstein algorithm.

Primitively the "user" could be used as an oracle to decide whether derivations are unique. If a violation is found attributes can be named but it would produce more derivations and possibly violations. It is not clear that such a process terminates by some given bound (an example of this will be discussed later). Beeri and Bernstein suggest an alternative solution--simply reject some inferences. We shall see that this may be necessary.

IV. A Partial Solution

We propose a classification of FDs, with which we can automatically detect many of the violations of uniqueness. We define three types of FDs: regular, injective and computable.

Armstrong's axioms are adapted to include the above types. A violation can be detected if two derivations of $X \twoheadrightarrow Y$ are found with different classifications. When two derivations of $X \twoheadrightarrow Y$ are of the same type then a violation can only be discovered by a finer classification. When both derivations are computable then a finer classification can be made by checking if they are the same computation. Unfortunately the general problem of equivalence is undecidable (depending on the language used). The situation is similar to program verification and most computations encountered will be simple enough to compare. The use of computable attributes in a data base is questionable and will be discussed later.

1. Regular FDs: These are those defined by Armstrong.

Example: $EMP^{\#} \twoheadrightarrow Dept^{\#}$; $Dept^{\#} \twoheadrightarrow MGR^{\#}$.

2. Injective (or One to One) FDs: This just means that the canonical function is always injective for each instance. We denote this by $X \dashrightarrow Y$.

Examples: $X \dashrightarrow X$, $SS^{\#} \dashrightarrow Passport^{\#}$.

3. Computable FDs: In this case the canonical function represents a real function of the attributes domains, other functions and the data base instance. We denote

this by $F:X \rightarrow Y$, where F represents the real function.

Notation: Lower case letters will denote the canonical names for non-computable FDs. Higher case letters will be reserved for computable FDs and general FDs (computable or non-computable) are denoted by Greek letters.

Examples:

A. $F: \text{SALARY, Number-of-Dependents} \rightarrow \text{Withholding-Tax}$

B. If A and B are attributes we may have $A+B=k$ a constant and $G:A \rightarrow B$ where $G=K-A$

C. $H: \text{Dept}^{\#} \rightarrow \text{Number-of-Employees}$

The algorithm for H is to count the number of employees in the department for each instance.

One may ask if computable attributes should be in a data base altogether. The answer is that in general they should not. If the computation is cheap it can be recalculated every time there is a query. Even if not the attribute should be virtual in the following sense:

1. The attribute should be attached to an appropriate relation- but not be considered in the relational schema and should not take part in questions of the various forms (i.e., not used for FDs).
2. Every time an update is made that affects the value of the virtual attribute in a tuple, it should be recalculated.

c. if $F: X \twoheadrightarrow Y$ and $(Y \cup Z \twoheadrightarrow W \text{ or } Y \cup Z \leq W)$ then $F^*: X \cup Z \twoheadrightarrow W$ where if f is the canonical name for $Y \cup Z \twoheadrightarrow W$ ($Y \cup Z \leq W$) then $F^*(X, Z) = f(F(X), Z)$.

d. if $(X \twoheadrightarrow Y \text{ or } X \leq Y)$ and $F: Y \cup Z \twoheadrightarrow W$ then $F^*: X \cup Z \twoheadrightarrow W$ where if f is the canonical name for $X \twoheadrightarrow Y$ ($X \leq Y$) then $F^*(X, Z) = F(f(X), Z)$.

e. if $F_1: X \twoheadrightarrow Y$ and $F_2: Y \cup Z \twoheadrightarrow W$ then $F^*: X \cup Z \twoheadrightarrow W$ where $F^*(X, Z) = F_2(F_1(X), Z)$.

It is obvious that the correct classification is given in each case.

We start with a set of user classified FDs. It can be assumed that there are no violations among the given FDs. Careful specification of the FDs will help prevent violations. We shall illustrate how the above classification of FDs can be used to detect violations by using the three examples given in Beeri and Bernstein [2].

In order to define computable functions we need a function manipulation language (we cannot use relations since we are only given FDs). Buneman and Frankel [3] have developed a function query language which would more than suffice for our purposes. Since its syntax is not commonly known we will develop only sufficient tools for our examples intuitively.

Non-computable FDs (denoted by lower case letters) are treated as atoms. We allow composition of functions (this was already used in the adapted axioms). Let $\alpha : A \rightarrow B$ (remember Greek letters represent both computable and non-computable FDs). α is realized in an extension of the data base as a partial function $\alpha : \text{dom}(A) \rightarrow \text{dom}(B)$ since $\text{dom}(A)$ may be infinite but only a finite number of values are realized in any extension. Let $\overline{\text{dom}(A)}$ represent the finite subset realized in a data base extension. Let $\alpha(A)$ represent the set $\{\alpha(a) \mid a \in \overline{\text{dom}(A)}\}$. For $b \in \overline{\text{dom}(B)}$, $\chi_{\#(A)=b}$ is the characteristic function of the predicate $\alpha(A)=b$ defined over $\text{dom}(A)$, i.e., $\chi_{\#(A)=b} : \text{dom}(A) \rightarrow \{0,1\}$ defined by:

$$\chi_{\#(A)=b}(a) = \begin{cases} 1 & \text{if } (a)=b \\ 0 & \text{otherwise} \end{cases}$$

We allow taking the sum of a set hence if $f : \text{Emp}^{\#} \rightarrow \text{Dept}^{\#}$ then we can define a computable function $F, F : \text{Dept}^{\#} \rightarrow \text{Number-of-Employees}$ by:

$$F(d) = \sum_{\text{dom}(\text{Emp}^{\#})} (\chi_{f(\text{Emp}^{\#})=d}) \text{ for } d \in \overline{\text{dom}(\text{Dept}^{\#})}.$$

Thus F would count the number of employees in a department.

Let $g : \text{Dept}^{\#} \rightarrow \text{Mgr}^{\#}$, we can define a computable $G : \text{Mgr}^{\#} \rightarrow \text{Number-of-Employees}$ by:

$$G(m) = \sum_{\text{dom}(\text{Dept}^{\#})} [(\chi_{g(\text{Dept}^{\#})=m}) * (F(\text{Dept}^{\#}))],$$

which computes the number of employees for a particular manager.

The above is just the embryo of a language but it is enough for our own purposes.

Example 1. We are given $f_1: \text{Dept}^\# \rightarrow \text{Mgr}^\#$, $f_2: \text{Emp}^\# \rightarrow \text{Dept}^\#$, $f_3: \text{Dept}^\#, \text{Floor} \rightarrow \text{Number-of-Employees}$, and $f_4: \text{Mgr}^\#, \text{Floor} \rightarrow \text{Number-of-Employees}$. f_3 is computable by:

$$f_3(a, f) = \sum_{\text{dom}(\text{Emp}^\#)} (\chi_{f_2(\text{Emp}^\#)=a, f})$$

f_4 is computable by:

$$f_4(m, f) = \sum_{\text{dom}(\text{Dept}^\#)} (\chi_{f_1(\text{Dept}^\#)=m}) * f_3(\text{Dept}^\#, f)$$

using $A_3(a)$ on $f_1: \text{Dept}^\# \rightarrow \text{Mgr}^\#$ and $f_4: \text{Mgr}^\#, \text{Floor} \rightarrow \text{Number-of-Employees}$ we derive $G: \text{Dept}^\#, \text{Floor} \rightarrow \text{Number-of-Employees}$ where,

$$G(a, f) = f_4(f_1(a), f) = \sum_{\text{dom}(\text{Dept}^\#)} (\chi_{f_1(\text{Dept}^\#)=f_1(a)}) * f_3(\text{Dept}^\#, f)$$

Clearly an algorithm could be defined which could decide G_{F_3} ; hence there are two derivations of $\text{Dept}^\#$, $\text{Floor} \rightarrow \text{Number-of-Employees}$ with different user intents. Beerl and Bernstein's solution to this violation is to change F_4 to:

$F_4: \text{mgr}^\#, \text{Floor} \rightarrow \text{Number-of-Employees-of-Manager}$

Remark. If we did not have t_2 (as is the case in the Beerl and Bernstein [2] example, F_3 would revert to a non-computable t_3 but F_4 would remain computable. G would be computable also and a violation would be detected because there were two derivations of $\text{Dept}^\#$, $\text{Floor} \rightarrow \text{Number-of-Employees}$ one regular and the other computable.

Example 2. Let $r_5: \text{Emp}^\# \rightarrow \text{Mgr}^\#$ and $f_6: \text{Mgr}^\# \leftarrow \text{Emp}^\#$. By $A_1(b)$ applied to f_6 we derive $g_1: \text{Emp}^\# \leftarrow \text{Mgr}^\#$ and this gives two derivations of $\text{Emp}^\# \rightarrow \text{Mgr}^\#$, one regular and the other injective. Here there is a violation. (We could have used transitivity on r_5, f_6 to get $g_2: \text{Emp}^\# \rightarrow \text{Emp}^\#$ as opposed to the derivation of $\text{Emp}^\# \leftarrow \text{Emp}^\#$ by $A_1(a)$.)

Beerl and Bernstein's solution to this violation is to change r_6 to $f_6: \text{Mgr}^\# \rightarrow \text{Emp}^\# \text{-of-Mgr}$. Unfortunately this leads to additional problems. Assume we have a hierarchy of managers (manager of managers, etc.); now would the manager of a manager be determined. If the manager is treated as a regular employee in $\text{Emp}^\# \rightarrow \text{Mgr}^\#$ an FD $\text{Emp}^\# \text{-or-Mgr} \leftarrow \text{Emp}^\#$ is needed which again causes a violation. Otherwise a $\text{Emp}^\# \text{-or-Mgr} \rightarrow \text{Mgr}^\# \text{-of-Mgr}$ is needed, and so on until the highest manager. This is analogous to a geneology data base with a Son, Father relation. In such a case

attributes for Grandfather, Greatgrandfather, etc. until Adam. This entails an explosion of attributes. The only reasonable solution is to outlaw problematic derivations and consider mgr -of-mgr etc. a virtual computable.

Example 3. Let $f_7: \text{Stock}^* \rightarrow \text{Store}^*$ and $f_8: \text{Stock}^*, \text{Store}^* \rightarrow \text{Qty}$. The "user intent" of f_7 is to map the Stock^* onto Store^* or the store that is in charge of ordering that item and f_8 maps Stock^* and Store^* of the store in which it is being sold into the quantity on hand. Using $A_3(a)$ we derive $g_3: \text{Stock}^* \rightarrow \text{Qty}$. Then $A_2(a)$ gives $g_4: \text{Stock}^*, \text{Store}^* \rightarrow \text{Qty}$. f_8 and g_4 represent two different intents of $\text{Stock}^*, \text{Store}^* \rightarrow \text{Qty}$ both classified regular. Hence they could not be distinguished by the classification method.

The above violation could be avoided by careful user definitions of FDs. The user should consider the range of the FD and if it does not include all the domain, a new attribute should be named. So $\text{Stock}^* \rightarrow \text{Store}$ doesn't include all store number in its range, only ordering store numbers.

IV. Discussion

It has been shown that checking the uniqueness assumption must be very time consuming. It is possible that certain limited types of sets of FDs cannot lead to violations but it is unlikely that these types could cover real situations.

Another problem encountered was a proliferation of attributes. Trying to satisfy the uniqueness assumption leads to many attribute names and it may make queries difficult for a user if he has to differentiate Ordering Room , Storing Room , Personnel Room , etc.

In addition the uniqueness assumption cannot contend with "natural loops" as in the $\text{Emp}^{\pm} \rightarrow \text{Mgr}^{\pm} \rightarrow \text{Emp}^{\pm}$ etc.

It is interesting to investigate what happens to these semantic violations in a regular relational data base using a normal query language.

Since the formulation of a query uses relation names the "derivation" depends on how the query is presented. We shall use example 2 and SEQUEL to illustrate. Assume we have two relations $E(\text{Emp}^{\pm}, \text{Mgr}^{\pm})$ and $M(\text{Mgr}^{\pm}, \text{Emp}^{\pm})$.

If we wanted the Emp^{\pm} of the manager of a given Emp^{\pm} , e, we could write:

```

Select      Emp±
From        M
Where       Mgr± =

              Select      Mgr±
              From        E
              Where       Emp± = e

```

We could also form the equation on Mgr getting EXM which are the triples, $(\text{Emp}^{\pm}, \text{Mgr}^{\pm}, \text{Emp}^{\pm})$, where the second Emp^{\pm} is the Emp^{\pm} of

the manager. The natural join of E and M would be null because of the violation. In a regular relational data base such violations may cause some problems but are not catastrophic. If the above violations were not detected, the Beerli and Bernstein algorithm could eliminate E or M as redundant. Hence the desirability of synthesizing relational data base schema should be considered.

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